

# Lecture 22

plan: 1) Finish matroid union ✓  
2) Ellipsoid

## Ellipsoid Algorithm

- general purpose convex opt. alg.
- poly time in many situations,  
e.g. linear programming.

(not necessarily strongly polynomial).

For  $\max \{c^T x : Ax \leq b\}$  can solve  
in  $\text{poly}(\langle A \rangle, \langle b \rangle)$   $\sum \log |a_{ij}|, \sum \log |b_i|$   
if  $a_{ij}, b_i \in \mathbb{Z}$ .

- Slow for LP in practice.

- Contrast w/ simplex which is fast in practice but not provably poly.
- May consequences for complexity of combinatorial opt problems.
- There is also interior point methods (Karmarkar '84) that solve LP in poly time, & fast in practice.  
but not as versatile for theory

# Consequences

Given convex set  $P \subseteq \mathbb{R}^n$ ,  
(e.g. a polyhedron), consider two  
problems:

- Separation (SEP):

Given  $y \in \mathbb{R}^n$ , decide if  
 $y \in P$ , if not return  
separating hyperplane i.e.  $c \in \mathbb{R}^n$   
s.t.  $c^\top y > \max \{c^\top x : x \in P\}$ .



## • optimization (OPT)

Given vector  $c \in \mathbb{R}^n$ , find  
find  $x$  maximizing  $c^T x$  on  $P$ .

## Examples

### • Linear programming:

$$P = \{x : Ax \leq b\}$$

how to solve SEP?



$P = \{x : a_i^T x \leq b_i\}$ , so just check  
foreach  $i$  if  $a_i^T x \leq b_i$ ;  
if not for some  $i$ ,  
output  $a_i^T x \leq b_i$  as separating hyperplane.

(efficient if  $A$  is part of  
the input)

- OPT for  $P = \{x : Ax \leq b\}$

is just LP.

$$\begin{aligned} & \max c^T x \\ & \text{subject to } Ax \leq b \end{aligned}$$

(SEP easy, OPT seems hard.)

- Matroid polytope:

$M = (E, \mathcal{I})$  matroid,

$$P = \text{conv}(\{1_S : S \in \mathcal{I}\})$$

matroid polytope

we know a face characterization.

Thm

$$P = \left\{ x \in \mathbb{R}^E : \begin{array}{l} x(S) \leq r(S) \quad \forall S \subseteq E \\ x_e \geq 0 \quad \forall e \in E \end{array} \right\}$$

However, exponentially many constraints!

Even if we can compute

Punk function  $f_M$ , SEP  
not obviously efficient!

- OPT for  $P$  is just  
greedy algorithm  
for the matroid.

$\text{OPT} = \max \text{ cost indep set.}$

( $\text{OPT}$  easy,  $\text{SEP}$  seems hard.)

- Matroid intersection polytope:

$\text{opt} ??? \text{SEP} ???$

- Amazing Result: Ellipsoid method & consequences in comb. opt.

Theorem (Grötschel, Lovasz, Schrijver '81) For a family  $P$  of convex bodies,

SEP for  $P$  is poly-time solvable



OPT for  $P$  is poly time solvable.

Proof idea:

⇒ Ellipsoid algorithm: can solve OPT using calls to SEP.

Reduces to using "polar" pt of  $P$ ; we won't cover.

- Actually, if  $P$  is "nondegenerate" enough, don't need SEP, just need **membership (MEM)** correction:

MEM! decide if  $x \in P$ .

Thm (GLS '88): Given ball of radius  $\epsilon$  contained in  $P$ , ball of radius  $R$  containing  $P$ , (and a MEM oracle to  $P$ ) can solve SEP with  $\text{poly}(\log(\frac{1}{\epsilon}), \log(R), n)$

→ Book

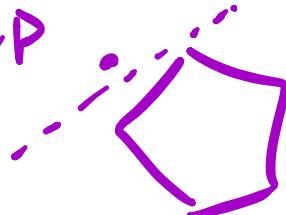
last time I didn't say!

if not given the  $\epsilon$ -ball, might never hit  $P$ !

Calls to MEM.

Actually, is about approximate versions of SEP & MEM.

Proof: not covered. SEP



## OPT vs. feasibility

- First we solve simpler problem:

FEAS (feasibility) Given SEP oracle for  $P$  find some  $x \in P$  or decide  $P = \emptyset$ .

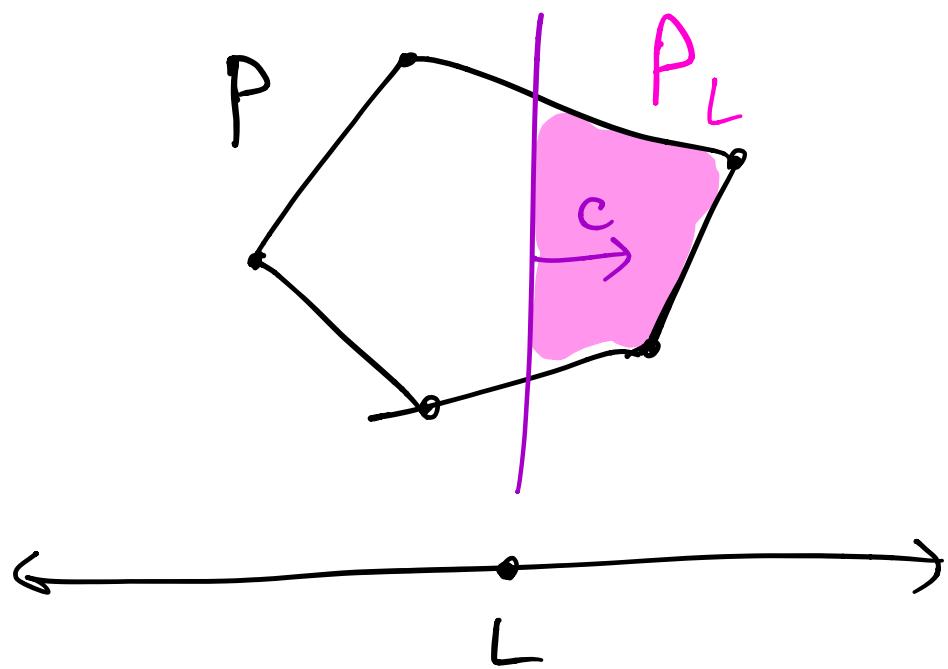
FEAS tells you if  $P$  empty or NOT.

- OPT reduces to FEAS:

### binary search :

$$\max \{c^T x : x \in P\} \geq L \text{ if}$$

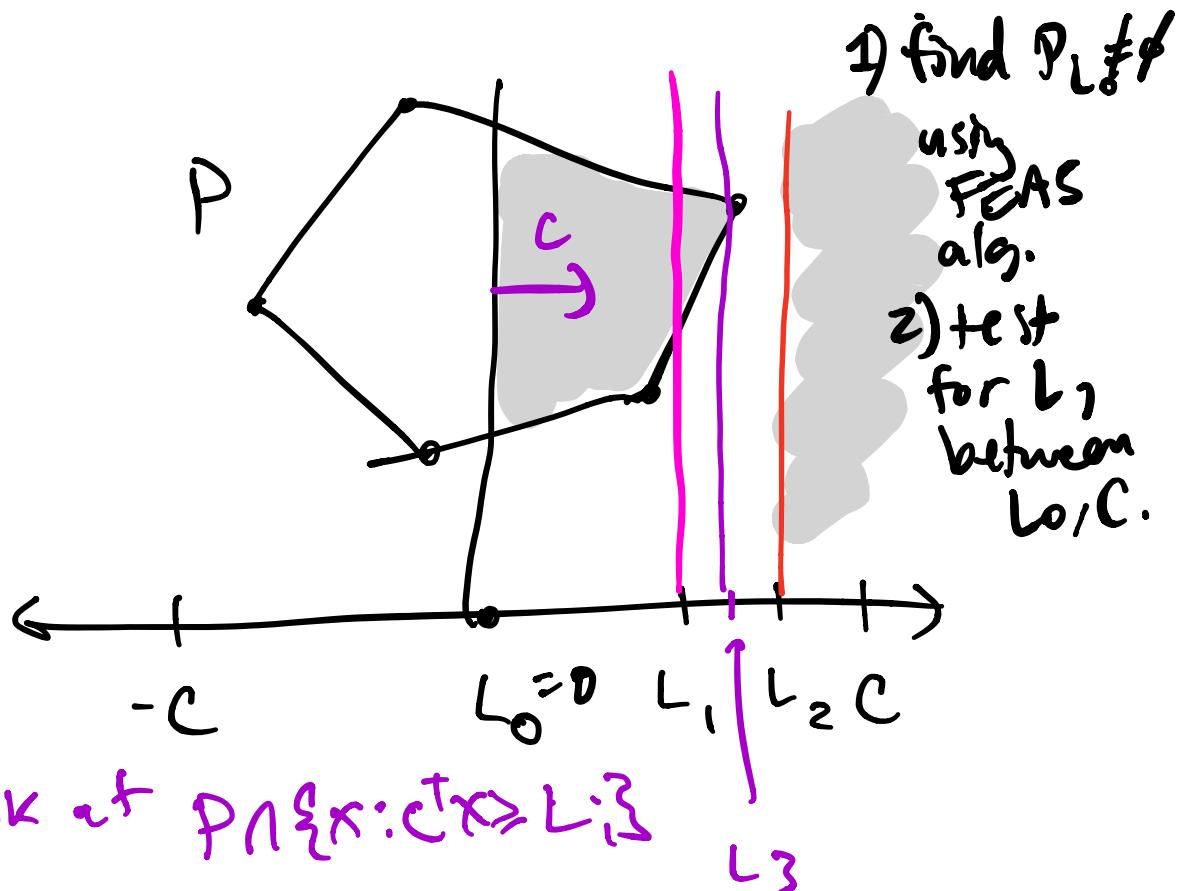
$$P_L := P \cap \{x : \hat{c}^T x \geq L\} \neq \emptyset.$$



- Given a-priori bound

$$-C \leq L \leq C$$

Binary search to find  $\max L : P_L \neq \emptyset$ .



- optimizes to  $\epsilon$ -precision  
in  $\log\left(\frac{2C}{\epsilon}\right)$  calls to FEAS.
- for  $L_f$ , can solve exact OPT  
with  $C, \frac{1}{\epsilon} \leq$  exponential in  
bitsize of  $A, b$ .

(Details later.)

(finally!)

## The Algorithm

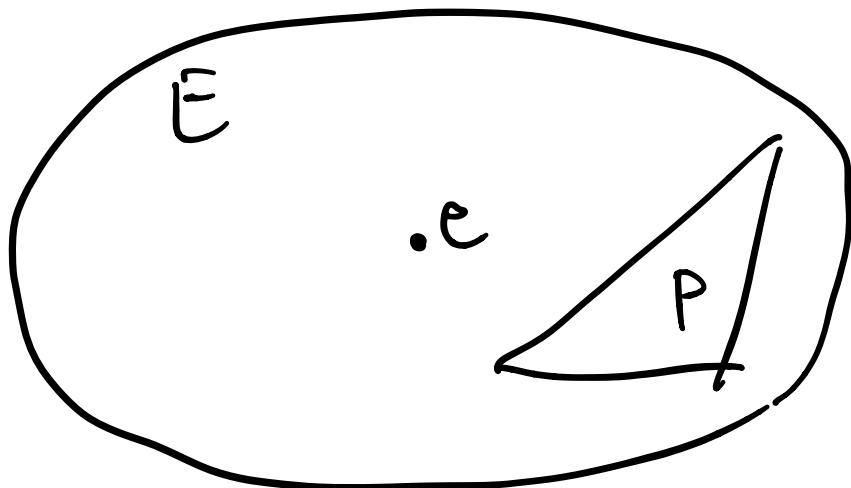
$P$   
or  
 $P = \emptyset$ .

- Solves FEAS in time  $\text{poly}(\log(\frac{1}{\epsilon}), \log R, n)$  assuming given ball  $B(x_0, R)$  of radius  $R$  containing  $P$ , and either  $P$  contains ball of radius  $\epsilon$  or  $P = \emptyset$ .
- $\epsilon, R$  dependence not a big deal:  
(& actually necessary).
  - ▷ For LP with  $P = \{x : Ax \leq b\}$   
 $\epsilon, R$  can be assumed exponential in bitsize of  $A, b$ .

using some tricks (we'll see  
these tricks for a  
special case.)

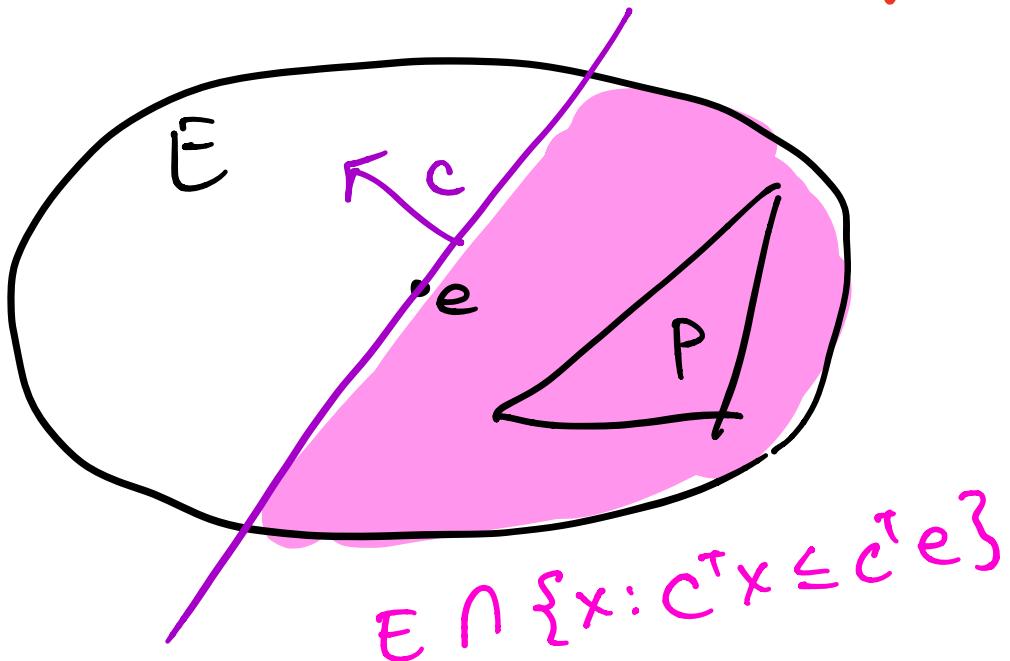
### Algorithm idea :

- Set  $E = E_0$ , ellipsoid guaranteed to contain  $P$ .  
(e.g.  $E_0 = \text{outer ball } B(0, R)$ ).

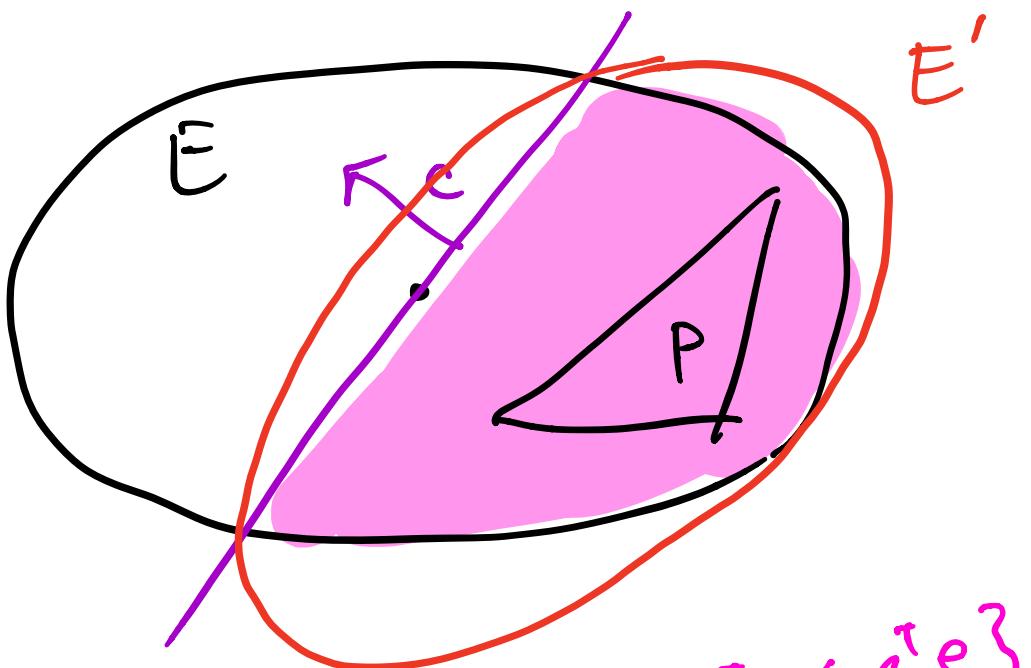


- While  $e \notin P$ : (if so, just return  $e$  & done).

▷ get separating hyperplane  
 ↗  $C^T x \leq d$  (valid for  $P$   
 but not for  $E$ ).  
 From  
 SEP  
 oracle. (actually, assume  $d = C^T e$ .  
 by translating the hyper  
 plane).



▷ Let  $E'$  "smaller ellipse" containing  $E \cap \{x : C^T x \leq C^T e\}$ .



$$E \cap \{x : C^T x \leq C^T e\}$$

(Can take  $E'$  to be minimum volume ellipsoid containing  $E \cap \{x : C^T x \leq C^T e\}$ , can find  $E'$  efficiently!)

▷ Set  $E \leftarrow E'$ .

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Runtime:

• Volume Lemma:

$$\text{vol}(E') \leq e^{-\frac{1}{2(n+1)}} \text{vol}(E).$$

- As  $E$  always contains  $P$ , alg. must terminate in

$$\leq 2(n+1) \log \left( \frac{\text{vol}(E_0)}{\text{vol}(P)} \right).$$

iterations.

(if  $P$  contains ball of radius  $\varepsilon$ ,  
contain ball of radius  $R$ ,  $\leq 2(n+1) \log \left( \left(\frac{R}{\varepsilon}\right)^n \right)$ )

$\leq 2(n+1)n \log\left(\frac{R}{\epsilon}\right)$ .  
If after  $2(n+1)n \log\left(\frac{R}{\epsilon}\right)$  iterations the algorithm hasn't terminated, output  $P = \emptyset$ .

### Issues:

- How to compute  $E'$
- Proof of volume lemma
- Bounding  $R, \epsilon$ .